

CS 5306
INFO 5306:
Crowdsourcing and
Human Computation

Lecture 17
10/31/17
Haym Hirsh



TECH

This AI came up with some hilarious Halloween costumes

By Lauren Tousignant

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137,571

Delta passengers busted having oral sex after meeting on plane



75,484





TECH

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To set up the Halloween generator, Shane entered 4,500 costume names into the network after crowdsourcing from the internet. The AI then took that data and tried to learn patterns and combine words.



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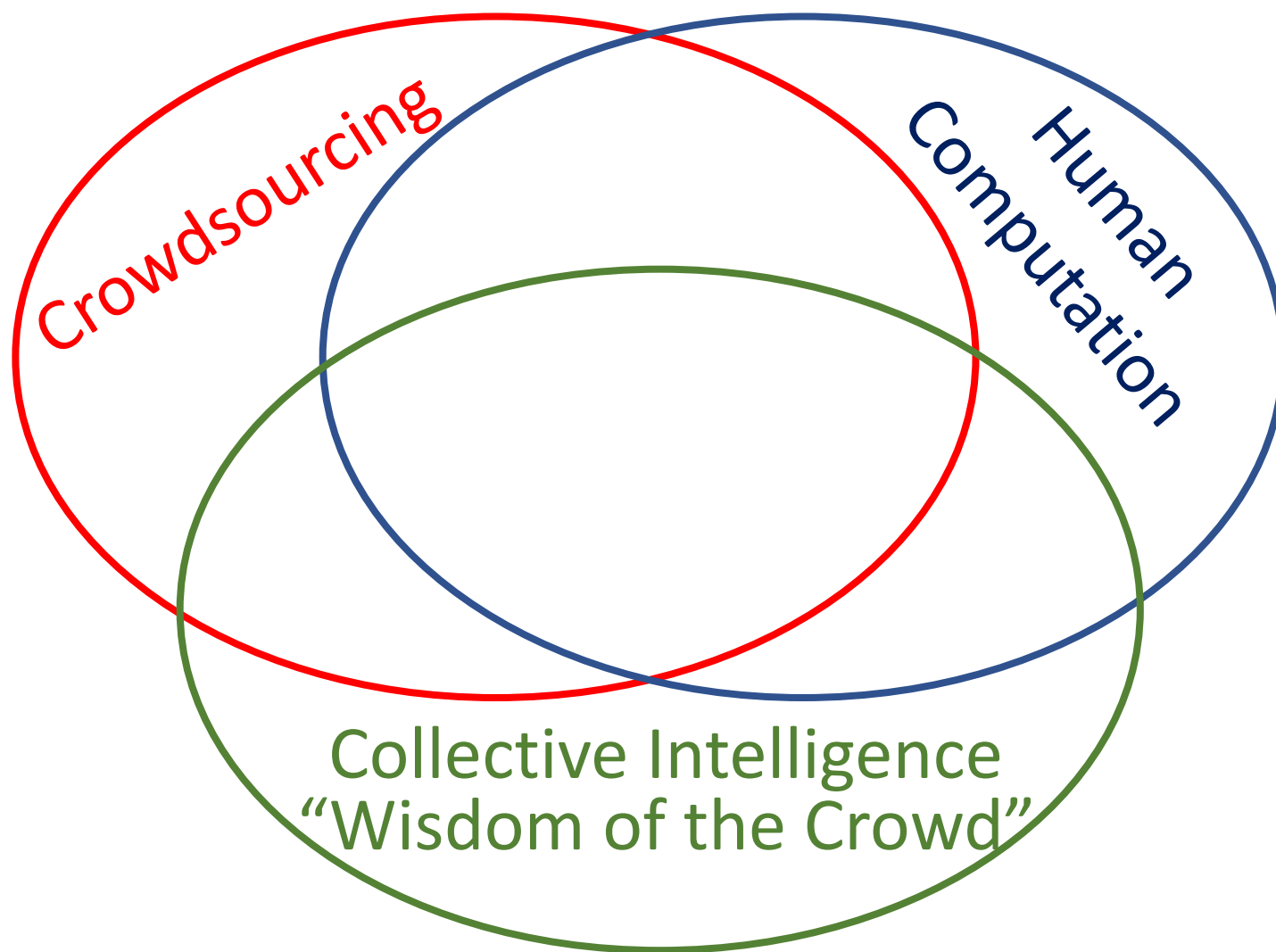


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Collective Intelligence

“Collective intelligence is a shared or group intelligence that emerges from the collaboration and competition of many individuals and appears in consensus decision making in bacteria, animals, humans and computer networks” - Wikipedia

Collective Intelligence

- Examples:
 - Animals:
 - Birds
 - Fish
 - Bees
 - Ants
 - Termites
 - Mold
 - Bacteria

Collective Intelligence

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People

Collective Intelligence

- Examples:
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People
Online

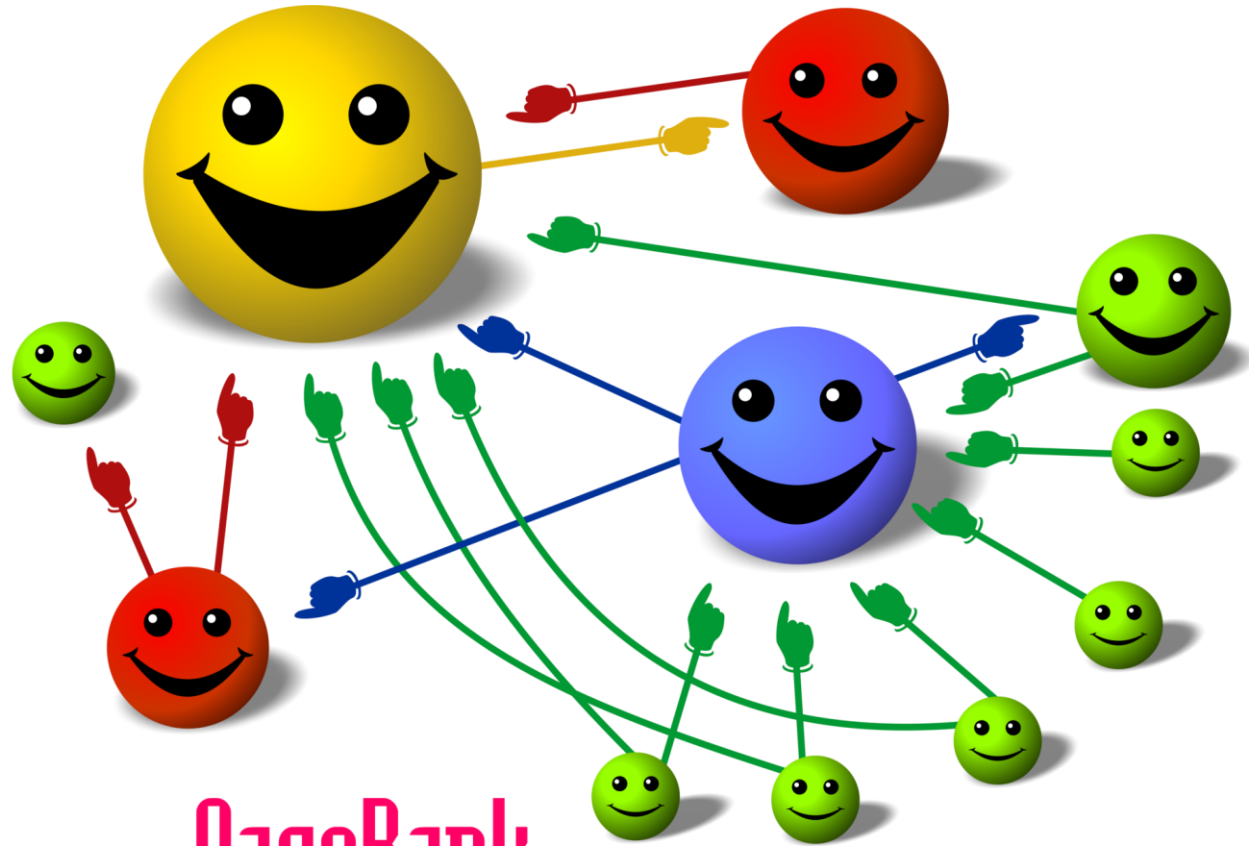
Google: Pagerank

- Pagerank: “The anatomy of a large-scale hypertextual Web search engine”, S. Brin and L. Page. In *Proceedings of the seventh international conference on the World Wide Web*, 1998.

Google: Pagerank

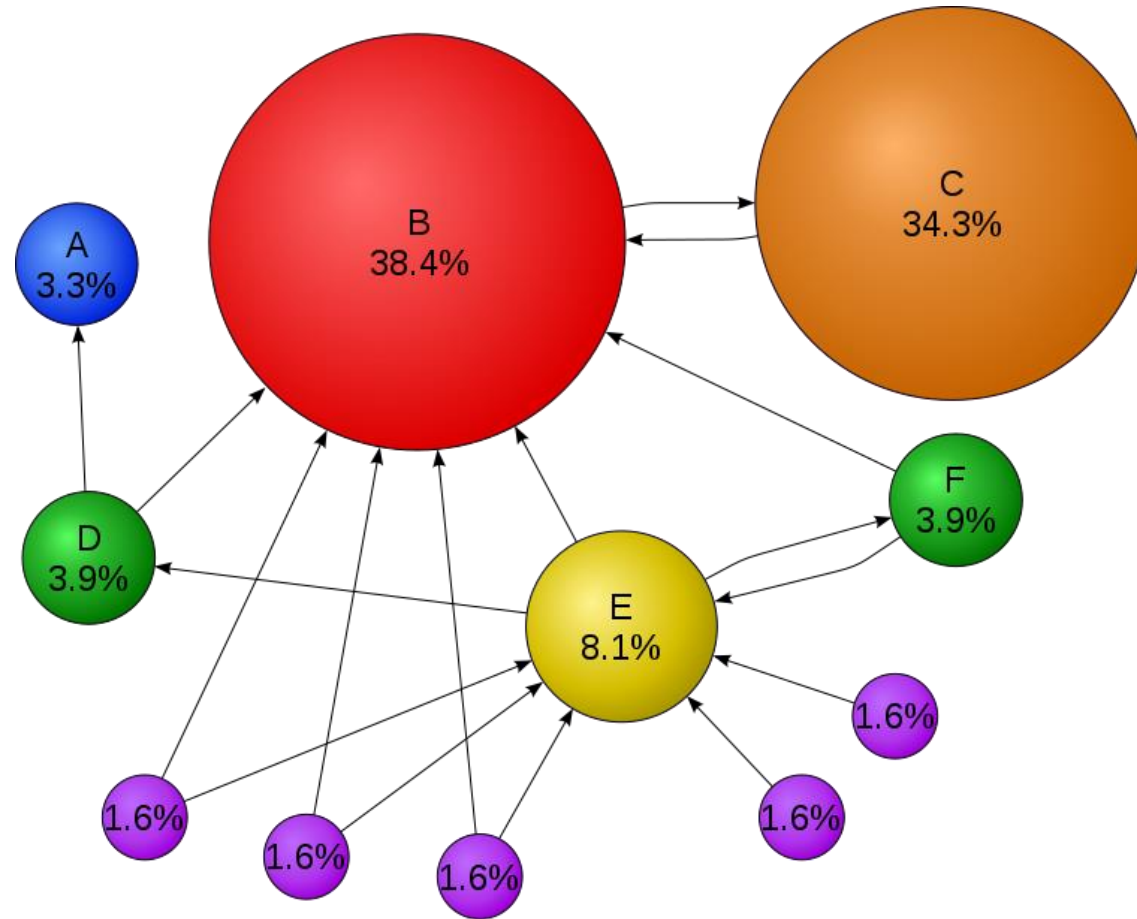
- Pagerank: “The anatomy of a large-scale hypertextual Web search engine”, S. Brin and L. Page. In *Proceedings of the seventh international conference on the World Wide Web*, 1998.
- HITS: “Authoritative sources in a hyperlinked environment”, Kleinberg JM. In *Proceedings of the ACM-SIAM Symposium on Discrete Algorithms* 1998.

Pagerank



PageRank

Pagerank



Pagerank

Imagine someone going randomly from page to page according to the popularity of a page (“random surfer model”).

For a page V_i :

$$S(V_i) = \sum_{j \in \text{in}(V_i)} \frac{S(V_j)}{|\text{out}(V_j)|}$$

$S(V_i)$ is proportional to the probability of landing on V_i .

Pagerank

- Problem: No way to leave a “sink”, so you’ll always end up in one

- Solution:
$$S(V_i) = (1-d) + d \sum_{j \in \text{In}(V_i)} \frac{S(V_j)}{|\text{Out}(V_j)|}$$

- d is the “damping factor” – probability of following a link versus taking a random jump somewhere else

Pagerank

- Solving
$$S(V_i) = (1-d) + d \sum_{j \in \text{In}(V_i)} \frac{S(V_j)}{|\text{Out}(V_j)|}$$
- Do a random walk
 - Start with equal probabilities for all nodes
 - Start at a random node
 - Take a step according to the probabilities of the edges (or make a random leap)
 - Repeat
- This converges to the correct S
- It can be solved using known method in computational linear algebra

Google Search Today

- Pagerank is one of 200 factors
 - Site specific
 - Page specific
 - User specific]